

DEVELOPMENT OF AGRICULTURE BASED ON DECISION TREE ALGORITHM

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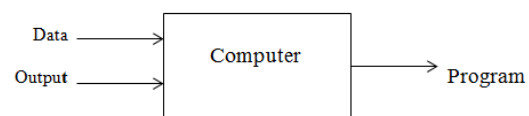
Abstract: Because of very quick swelling of population in world, severe transformation in supply of food, land resources constraint gives burden for agriculture stream in leading than ever before. In agriculture production lot of element plays vital function in agriculture output. In this study proposes the decision tree algorithm to investigate the different elements which affects and demanding to establish the root component of production control. In this article describes the fundamental theory of decision tree technique which is used to forecast or predict and classify the agriculture data. For all class of agriculture data this decision tree method is utilized for prediction and classification.

Keywords- Machine learning, Decision tree, Classification, Agriculture.

1.INTRODUCTION

Machine learning is a branch of science that arrangements with programming the frameworks such that they consequently learn and enhance with understanding. Here, learning implies perceiving and understanding the information and settling on savvy choices in view of the provided information [1]. It is extremely hard to provide food every one of the choices in view of every single conceivable info. To beat this issue, calculations are produced. These calculations construct information from particular information and past involvement with the standards of measurements, likelihood hypothesis, rationale, combinatorial improvement, seek, fortification learning, and control hypothesis [2]. Machine learning is a huge zone and it is very past the extent of this instructional exercise to cover every one of its highlights.

There are a few approaches to actualize machine learning systems, however the most generally utilized ones are managed and unsupervised learning [3],[4]. In conventional programming data and program is keep running on the PC to deliver the yield. Be that as it may, in Machine learning as appeared in fig 1 information and yield is keep running on the PC to make a program [5]. This program can be utilized as a part of customary programming.



|Figure 1 Machine learning

There are four types of machine learning they are

1. Supervised learning

2. Unsupervised learning
3. Semi-supervised learning
4. Reinforcement learning

If the training data includes desired outputs then it is called as supervised learning or inductive learning but in unsupervised learning the Training data does not include desired outputs [16]. If the training data includes a few desired outputs it is known as Semi-supervised learning [6].

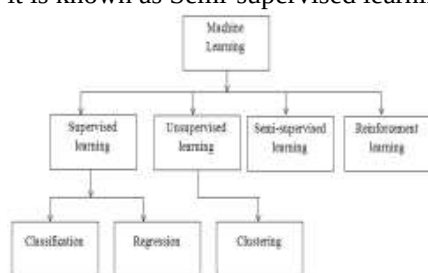


Figure 2 Types of machine learning

Now a-days for extremely large applications data or information which stored in database can be used.. This dangerous development in information and database has created an earnest requirement for new methods and instruments that can wisely naturally change the prepared information into valuable data and learning. Henceforth information mining has turned into an examination are with expanding significance. Characterization and forecast are two types of information examination that can be utilized to remove models portraying essential information classis or to anticipate future information patterns. Such examination can assist furnish us with better comprehension of the information on the loose [7]. Though arrangement predicts straight out (discrete, unordered) marks, expectation models persistent esteemed capacities. Numerous grouping and forecast techniques have been proposed by scientist in machine learning design acknowledgment and insights. Most calculations are memory

occupant, ordinarily accepting a little information measure [8] . Late information mining research has based on such work, creating adaptable arrangement and expectation methods fit for taking care of expansive plate inhabitant data. There are arrangements of information mining strategies, for example, visit design mining, classification, relapse, grouping, affiliation administer mining and numerous more, but out of these classification is as often as possible utilized and no more. In classification, the model is prepared which depicts and differentiate different information classes to foresee the classes whose names are not known. The classification can be performed with different calculations, for example, neural systems, choice trees, relapse, and so on. Because of the significance essential of choice tree for huge informational indexes, in this examination choice tree approach has been utilized.

2. DECISION TREE

A tree can be learned by part the source set into subsets in light of a quality esteem test. This procedure is rehashed on each determined subset in a recursive way called recursive apportioning. The recursion is finished when the subset at a hub has all a similar estimation of the objective variable, or while part never again enhances the forecasts [9] . This procedure of best down acceptance of choice trees (TDIDT) is a case of an eager calculation, and it is by a long shot the most widely recognized technique for taking in choice trees from information. The choice tree comprises of three basics, root hub, inside hub and leaf hub [10] . Top most central is the root hub. Leaf hub is the terminal major of the structure and the hubs in the middle of is known as the inside hub. Each inward hub indicates test on a trait, each branch speaks to a result of the test, and each leaf hub holds a class mark. Different choice tree calculations are

utilized as a part of order like ID3, J48, CART, C5.0, SLIQ, SPRINT, irregular backwoods, arbitrary tree, and so forth. In this work following tree calculations are taken for examination.

ID3 (Iterative Dichotomiser 3) decision tree algorithm is produced by Quinlan. The essential thought of ID3 calculation is to develop the choice tree by utilizing a best down, voracious inquiry through the offered sets to test each trait at each tree hub. In the choice tree technique, data pick up approach is for the most part used to decide reasonable property for every hub of a produced choice tree. Along these lines, we can choose the characteristic with the most astounding data pick up (entropy lessening in the level of greatest) as the test property of current hub. Along these lines, the data expected to characterize the preparation test subset got from later on dividing will be the littlest. In other words, the utilization of this property to parcel the example set contained in current hub will make the blend level of various kinds for all produced test subsets decrease to a base [11]. Along these lines, the utilization of such a data hypothesis approach will successfully decrease the required partitioning number of protest arrangement. ID3 utilizes just unmitigated ascribes to fabricate a tree demonstrate. This calculation does not create more precise results if the clamor is available in the informational index. Keeping in mind the end goal to get more exact outcomes the compelling pre-preparing is done before show is constructed utilizing ID3.

J48 is a propel rendition of ID3, it chooses target estimation of another test information as for various characteristic benefits of preparing information [3]. The inner hubs of a choice tree are meant by various properties while the branches tell the conceivable

estimations of these qualities. The inside hubs tell the reliant variable esteems.

CART, its full form is Classification And Regression Tree (CART) and it was at first proposed by Breiman et al. which was double tree otherwise called (HODA). It is non parametric choice tree. It produces relapse or grouping tree relies upon subordinate variable's compose either numeric or straight out individually. Here the importance of parallel means the hub in a choice tree has two out word branches i.e. gatherings. Gini record is utilized as a part of CART as highlight choice measure. The trait with biggest gini list is utilized to part the records. Truck handles both straight out what's more, numerical esteems alongside missing quality esteems too. It utilizes costcomplexity pruning and furthermore create relapse trees. It assembles the two groupings and relapse trees. It can be executed serially from Hunt's calculation. It additionally utilizes relapse investigation utilizing regressin trees (S.Anupama et al,2011). Over a given timeframe and the arrangement of indicator factors relapse investigation highlight conjectures a needy variable. It gives high characterization and expectation exactness.

C5.0 is an augmentation of C4.5 which was at first got from ID3. It is connected on enormous informational collections. It is substantially speedier and memory effective than C4.5. It parts the examples in view of the most extreme data pick up. The example subset that is get from the previous part will be part a while later. This is consistent process until the point when the example subset can not be additionally part. The characteristics/highlights which have less commitment will be rejected. One noteworthy preferred standpoint of C5.0 is it handles multi esteem properties and missing qualities from the informational index

3.DECISION TREE WORKING METHODOLOGY

1. First choose the training experience or data
2. Choose target function that is to be learned.
3. Choose how to represent the target function.
4. Finally choose the learning algorithm to enter the target function.

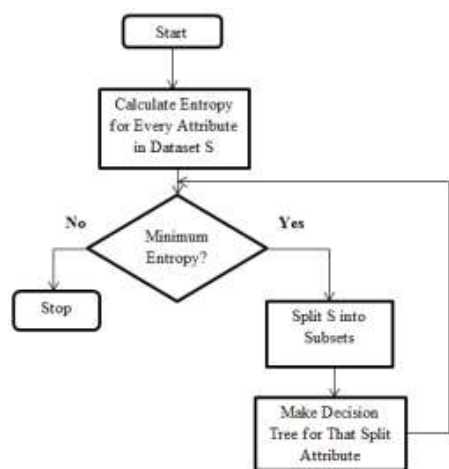


Figure 3 Working of decision tree

For the most part the classification or regression is the grouping of the tasks as takes after:

1. Set up the preparation informational index utilizing pre-handling on the raw data.
2. Class attribute and classes are identified.
3. Recognize helpful properties for classification (Relevance investigation).
4. Take in a model utilizing preparing cases in Training set.
5. Utilize the model to characterize the obscure information tests.

TABLE 1 Performance based comparisons of different decision tree algorithms

Algorithms	Procedure	Measure	Pruning
ID3	Top-down approach	Entropy, Information Gain	Pre-pruning
C4.5	Top-down approach	Entropy, Information Gain ratio, Split info	Pre-pruning
C5.0	Top-down approach	Entropy, Information Gain ratio, Split info	Post-pruning
CART	Binary decision tree construction	Gini diversity index	Post-pruning based on complexity

TABLE 2 Comparisons between different Decision Tree Algorithms

Algorithms	ID3	C4.5	C5.0	CART
Processing speed	Slow	Better	Faster	Average
Types of data	Categorical	Categorical, Numerical	Categorical, dates, time, Numerical	Numerical, Nominal
Boosting	Not Allowed	Not Allowed	Allowed	Allowed
Pruning	No	Pre-pruning	Post-pruning	Pre-pruning

Measure	Entropy, Information Gain	Information Gain ratio, Split info	Information Gain ratio, Split info	Gini diversity index
Missing values	Not Supported	Not Supported	Supported	Supported

3.1 Issues in decision trees

While modelling the decision tree over-fitting is one of the realistic issues faced. When constructing the model with many branches over-fitting happens because of the presence of outliers and missing information in the dataset. Consequently it will decrease the training set error and increase the test set error results in test set accuracy of the model decreases. In order to overcome these issues two approaches can be introduced.

1. Pre-pruning
2. Post-pruning

Pre-pruning:

Pruning is the process of reducing the size of decision tree in order to minimize the misclassification error rate.

After checking some performance measures, the process of cut down or halting the sub-branch of decision tree at some node is called pre-pruning. It is also called as early stopping because in this technique it stops the growing of tree prematurely.

Post-pruning

This method allows the decision tree to its entirety. After that it trim or removes the nodes of decision tree in a bottom-up procedure. It can be done by replacing the node with leaf.

3.2 Advantages:

1. Interpretable initially
2. Suitable for taking care of both all out and quantitative estimates.
3. Universal for taking care of both classification and regression issues
4. Easy to understand.
5. Easy to produce rules

3.3 Disadvantages:

1. May experience the ill effects of overfitting
2. Classifies by rectangular partitioning
3. Does not deal with non-numeric information
4. Can be very substantial so pruning is important
5. Decision trees can be unsteady.
6. It can be hard to control the measure of the tree.

3.4 Applications:

- **Web search:** positioning page in light of what you are well on the way to tap on.
- **Computational biology:** balanced outline sedates in the PC in view of past tests.
- **Finance:** choose who to send what charge card offers to. Assessment of hazard on layaway offers. Instructions to choose where to contribute cash.
- **E-trade:** Predicting client beat. Regardless of whether an exchange is false.
- **Space exploration:** space tests and radio stargazing.
- **Robotics:** how to deal with vulnerability in new situations. Self-ruling. Self-driving auto.
- **Information extraction:** Ask inquiries over databases over the web.

4. RESULTS AND DISCUSSION

The experimental data are from the total output value of an annual agriculture production, there lists data of 10 provinces and cities' (such as, Delhi, Chandigarh, Nasik, Kanpur, Surat, and so on) labors in rural area, land area and total yield of agriculture. The attributes used is shown in Table 3.

TABLE 3 Agriculture production value

City	Rural labor	Arable land Area	Gross Agriculture production
Delhi	67.7	399.5	176.58
Srinagar	79.4	426.1	156.17
Chandigarh	1635.5	6517.3	1505.94
Surat	639.9	3645.1	359.15
Chennai	512.4	5491.4	534.39
Kanpur	633.0	3389.7	969.79
Jaipur	517.0	3953.2	666.47
Hyderabad	760.3	8995.3	736.34
Bhopal	76.3	290.0	206.78
Nasik	1531.5	4448.3	1849.18

Use decision tree to classify the attributes contains city,labours in rural area, land area and yield of agriculture and the results are listed in Table 4. To carry out the generalization on output of agriculture value by decision tree method consider the following allocation. For Rural labor attribute output 1 indicates few, 2 indicates medium and 3 indicates much. For land area attribute output 1 indicates small, 2 indicates medium and 3 indicates large. For land area attribute output 1 indicates low, 2 indicates medium and 3 indicates high.

The details of total yield of agriculture is given in Table 5

TABLE 4 Result by Decision Tree method

City	Rural labor Class	Arable land Area Class	Gross Agriculture Production Class
Delhi	1	1	1
Srinagar	1	1	1
Chandigarh	2	3	2
Surat	1	2	1
Chennai	1	3	1
Kanpur	1	2	2
Jaipur	1	2	1
Hyderabad	1	3	1
Bhopal	1	1	1
Nasik	2	2	3

TABLE 5 Output value of attributes after generalization

City	Rural labor Class	Arable land Area Class	Gross Agriculture Production Class
Delhi	Few	Small	Low
Srinagar	Few	Small	Low
Chandigarh	Medium	Large	Medium
Surat	Few	Medium	Low
Chennai	Few	Large	Low
Kanpur	Few	Medium	Medium
Jaipur	Few	Medium	Low
Hyderabad	Few	Large	Low
Bhopal	Few	Small	Low
Nasik	Medium	Medium	High

Out of these analysis, the decision tree of various such as land area, labours in rural areas and total yield of agriculture, From this it is clear that total value of Agricultural yield is mainly depends on number of rural

labor. If the number of labors of various cities is high then the agriculture yield also high.

5. CONCLUSION

In this article, decision tree algorithm is utilized to classify or predict the agriculture data. Although few data files contains lost of data. This proposed method capable to handle both finished and unfinished data. As per our analysis, the efficiency of the decision tree is completely trust in entropy, information gain for classification techniques and standard deviation, standard deviation reduction for prediction techniques. By utilizing the decision tree properly different types of work has been completed. This analysis gives some primary key knowledge to the student and researcher about benefits, difficulties and issues of decision tree algorithm.

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